
Errata to Introduction to Derivative-Free Optimization (05/17/2015)

A. R. Conn, K. Scheinberg, and L. N. Vicente, Introduction to Derivative-Free Optimization, MPS-SIAM Book Series on Optimization, SIAM, Philadelphia, 2009

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1. Page 9, Line 17: “The unique minimizer is at $(0, 0)$ (see Figure 1.1).”

Corrected version: The contours of this function are plotted in Figure 1.1. The unique minimizer of the underlying smooth function $(10x_1^2 + 2x_2^2 + 4x_1x_2)$ is at $(0, 0)$.

2. Page 10, Line -4 (from below): “The Nelder–Mead method failed for the initial point”

Corrected version: The Nelder-Mead method is trapped at a spurious minimizer for the initial point

3. Lines 4–19 in Page 29 including the footnote should be changed to the following (*note, however, that the precise statement of Theorem 2.13 as it stands in the book would be valid if the regression model is of the form $m(y) = f(y^0) + (y - y^0)^\top g$ and that such a g would then coincide with the regression simplex gradient of Page 33*):

The proof of the bounds, when $y^0 = 0$, can follow approximately the same steps as the proof for the linear interpolation case. Note that one can assume $y^0 = 0$ without loss of generality (see the argument at the end of the proof of Theorem 3.16). Considering $y^0 = 0$, one has

$$M = \begin{bmatrix} 1 & 0 \\ e & L \end{bmatrix} \quad \text{and} \quad \hat{M} = \begin{bmatrix} 1 & 0 \\ e & \hat{L} \end{bmatrix},$$

where e is the vector of ones of dimension p . Now, note that

$$M \begin{bmatrix} f(y^0) \\ \nabla f(y^0) \end{bmatrix} - f(Y) = r,$$

with $f(Y) = (f(y^0), f(y^1), \dots, f(y^p))^\top$ and $|r_i| \leq (\nu/2)\Delta^2$, $i = 0, \dots, p$. Thus, one obtains³

$$\begin{bmatrix} f(y^0) \\ \nabla f(y^0) \end{bmatrix} - M^\dagger f(Y) = \begin{bmatrix} f(y^0) \\ \nabla f(y^0) \end{bmatrix} - \begin{bmatrix} c \\ g \end{bmatrix} = M^\dagger r.$$

Noting that (I is the identity matrix of order p)

$$M^\dagger = \begin{bmatrix} 1 & 0 \\ 0 & (1/\Delta)I \end{bmatrix} \hat{M}^\dagger, \tag{2.9}$$

³ A^\dagger denotes the Moore-Penrose generalized inverse of a matrix A , which can be expressed by the singular value decomposition of A for any real or complex matrix A . In the current context where M is full column rank, we obtain the left inverse $M^\dagger = (M^\top M)^{-1} M^\top$.

we can then state the bounds in a format similar to the linear interpolation case.

Theorem 2.13. *Let Assumption 2.2 hold. The gradient of the linear regression model satisfies, for all points y in $B(y^0; \Delta)$, an error bound of the form*

$$\|\nabla f(y) - \nabla m(y)\| \leq \kappa_{eg} \Delta,$$

where $\kappa_{eg} = \nu(1 + p^{\frac{1}{2}}\|\hat{M}^\dagger\|/2)$ and \hat{M} is the scaled version of M given above.

The linear regression model satisfies, for all points y in $B(y^0; \Delta)$, an error bound of the form

$$|f(y) - m(y)| \leq \kappa_{ef} \Delta^2,$$

where $\kappa_{ef} = 2\kappa_{eg} + \nu/2$.

4. Page 33, Lines 13–14: “Again, one points out that a simplex gradient defined in this way is the gradient g of the linear regression model $m(x) = c + g^\top x$.”

Corrected version: Again, one points out that a simplex gradient defined in this way is the gradient g of a linear regression model, if written as $m(x) = f(y^0) + g^\top(x - y^0)$.

5. Page 33, Line 18: “It is then obvious”

Corrected version: It can be proved using arguments already seen

6. Page 34, Line 9: “10. From (2.9), conclude the proof of Theorem 2.13.”

Corrected version (note that (2.9) has changed above): “10. Conclude the proof of Theorem 2.13 by first showing (2.9).”

7. Page 43, Lines 7–8: “Figures 3.1–3.4 show several sets of six points in B — the ‘squared’ ball of radius $1/2$ around the point $(0.5, 0.5)$ in \mathbb{R}^2 .”

Corrected version: “Figures 3.1–3.4 show several sets of six points in a neighborhood of B , the (Euclidean) ball of radius $1/2$ centered at the point $(0.5, 0.5)$ in \mathbb{R}^2 .”

8. Page 43, Line 2 of the caption of Figure 3.1: “and $\Lambda = 440$.”

Corrected version: “and $\Lambda = 294$.”

9. Page 44, Line 2 of the caption of Figure 3.2: “and $\Lambda = 21296$.”

Corrected version: “and $\Lambda = 5324$.”

10. Page 44, Line 3 of the caption of Figure 3.3: “and $\Lambda = 524982$.”

Corrected version: “and $\Lambda = 492624$.”

11. Page 69 (last 3 lines): What should be there is instead:

$$\begin{aligned}\kappa_{eh} &= \nu_2 + \sqrt{2\bar{p}^{\frac{1}{2}}}\nu_2/2\|\hat{\Sigma}^{-1}\|, \\ \kappa_{eg} &= \nu_2 + (n^{\frac{1}{2}} + \sqrt{2\bar{p}^{\frac{1}{2}}})/2\nu_2\|\hat{\Sigma}^{-1}\|, \\ \kappa_{ef} &= \nu_2/2 + (1/2 + n^{\frac{1}{2}}/2 + \sqrt{2\bar{p}^{\frac{1}{2}}}/4)\nu_2\|\hat{\Sigma}^{-1}\|,\end{aligned}$$

with $\bar{p} = n(n+1)/2$.

12. Page 81, Line 6: “of the Hessian of $m(x)$.”

Corrected version: “of the upper or lower triangular part of the Hessian of $m(x)$.”

13. Page 98, Line -4: “with $\Lambda = 21296$.”

Corrected version: “with $\Lambda = 5324$.”

14. Page 98, Line -2 and -1: The sentence “This involves scaling the points by a factor of two, hence the Λ constant is reduced by a factor of 4 to $\Lambda = 5324$.” should be removed.

15. The sets Y_1 and Y_2 in Page 99 should be changed to

$$Y_1 = \begin{bmatrix} -0.98 & -0.96 \\ -0.96 & -0.98 \\ 0 & 0 \\ 0.98 & 0.96 \\ 0.96 & 0.98 \\ 0.707 & -0.707 \end{bmatrix}, \quad Y_2 = \begin{bmatrix} -0.848 & 0.528 \\ -0.96 & -0.98 \\ 0 & 0 \\ 0.98 & 0.96 \\ 0.96 & 0.98 \\ 0.707 & -0.707 \end{bmatrix}$$

16. Page 119, Line 30: “the number of positive bases is required to be finite.”

Corrected version: the number of positive bases (from which the poll directions are extracted; see Sections 7.5 and 7.6) is required to be finite.

17. Page 232, Line 17: “for $i, j \in \{0, \dots, p\}$ ”

Corrected version: for $i \in \{0, \dots, p\}$ and $j \in \{0, \dots, q\}$

18. Page 245, Line 14: “the following is a set of positive generators”

Corrected version: the following includes a set of positive generators

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